UNITED STATES PATENT APPLICATION

OF

ALEXANDER KELLER

FOR

SYSTEM AND METHOD FOR GENERATING PIXEL VALUES FOR PIXELS IN AN IMAGE USING

STRICTLY DETERMINISTIC METHODOLOGIES FOR GENERATING SAMPLE POINTS

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By _____

Richard A. Jordan

1 **INCORPORATION BY REFERENCE**

2 U. S. Patent Application Serial No. 08/880,418, filed June 23, 1997, in the names of Martin

3 Grabenstein, et al., entitled "System And Method For Generating Pixel Values For Pixels In An

4 Image Using Strictly Deterministic Methodologies For Generating Sample Points," (hereinafter

5 referred to as the Grabenstein application) assigned to the assignee of this application, incorporated

6 by reference.

FIELD OF THE INVENTION

The invention relates generally to the field of computer graphics, and more particularly to
systems and methods for generating pixel values for pixels in the image.

BACKGROUND OF THE INVENTION

In computer graphics, a computer is used to generate digital data that represents the

12 projection of surfaces of objects in, for example, a three-dimensional scene, illuminated by one or

13 more light sources, onto a two-dimensional image plane, to simulate the recording of the scene by,

14 for example, a camera. The camera may include a lens for projecting the image of the scene onto

15 the image plane, or it may comprise a pinhole camera in which case no lens is used. The two-

16 dimensional image is in the form of an array of picture elements (which are variable termed "pixels"

17 or "pels"), and the digital data generated for each pixel represents the color and luminance of the

18 scene as projected onto the image plane at the point of the respective pixel in the image plane. The

19 surfaces of the objects may have any of a number of characteristics, including shape, color,

20 specularity, texture, and so forth, which are preferably rendered in the image as closely as possible,

21 to provide a realistic-looking image.

22 Generally, the contributions of the light reflected from the various points in the scene to the

23 pixel value representing the color and intensity of a particular pixel are expressed in the form of the

24 one or more integrals of relatively complicated functions. Since the integrals used in computer

1   graphics generally will not have a closed-form solution, numerical methods must be used to evaluate

2   them and thereby generate the pixel value. Typically, a conventional "Monte Carlo" method has been

3   used in computer graphics to numerically evaluate the integrals. Generally, in the Monte Carlo

4   method, to evaluate an integral

5

$$\langle f \rangle \;=\; \int_0^1 f(x)\,dx \tag{1}$$

6   where f(x) is a real function on the real numerical interval from zero to 1, inclusive, first a number

7   "N" statistically-independent random numbers xi , i=1, ..., N, are generated over the interval. The

8   random numbers xi are used as sample points for which sample values f(xi) are generated for the

9   function f(x), and an estimate $\bar{f}$ for the integral is generated as

10

$$\langle f \rangle \;\approx\; \bar{f} \;=\; \frac{1}{N}\sum_{i=1}^{N} f(x_i) \tag{2}$$

11   As the number of random numbers used in generating the sample points f(xi) increases, the value

12   of the estimate $\bar{f}$ will converge toward the actual value of the integral $\langle f \rangle$. Generally, the

13   distribution of estimate values that will be generated for various values of "N," that is, for various

14   numbers of samples, of being normal distributed around the actual value with a standard deviation

15   σ which can be estimated by

16

$$\sigma \;=\; \sqrt{\frac{1}{N-1}\left(\overline{f^2} - \bar{f}^2\right)} \tag{3}$$

17   if the values x_i used to generate the sample values $f(x_i)$ are statistically independent, that is, if the

18   values x_i are truly generated at random.

19   Generally, it has been believed that random methodologies like the Monte Carlo method are

20   necessary to ensure that undesirable artifacts, such as Moiré patterns and aliasing and the like, which

1   are not in the scene, will not be generated in the generated image. However, several problems arise

2   from use of the Monte Carlo method in computer graphics. First, since the sample points x_i used in

3   the Monte Carlo method are randomly distributed, they may clump in various regions over the

4   interval over which the integral is to be evaluated. Accordingly, depending on the set of random

5   numbers which are generated, in the Monte Carlo method for significant portions of the interval

6   there may be no sample points x_i for which sample values $f(x_i)$ are generated. In that case, the error

7   can become quite large. In the context of generating a pixel value in computer graphics, the pixel

8   value that is actually generated using the Monte Carlo method may not reflect some elements which

9   might otherwise be reflected if the sample points x_i were guaranteed to be more evenly distributed

10   over the interval. This problem can be alleviated somewhat by dividing the interval into a plurality

11   of sub-intervals, but it is generally difficult to determine a priori the number of sub-intervals into

12   which the interval should be divided, and, in addition, in a multi-dimensional integration region

13   (which would actually be used in computer graphics, instead of the one-dimensional interval

14   described here) the partitioning of the region into sub-regions of equal size can be quite complicated.

15   In addition, since the method makes use of random numbers, the error $\left| \bar{f} - \langle f \rangle \right|$ (where $|x|$

16   represents the absolute value of the value "x") between the estimate value $\bar{f}$ and actual value $\langle f \rangle$

17   is probabilistic, and, since the error values for various large values of "N" are close to normal

18   distribution around the actual value $\langle f \rangle$, only sixty-eight percent of the estimate values $\bar{f}$ that

19   might be generated are guaranteed to lie within one standard deviation of the actual value $\langle f \rangle$.

20   Furthermore, as is clear from equation (3), the standard deviation σ decreases with increasing

21   numbers of samples N, proportional to the reciprocal of square root of "N" (that is, $\frac{1}{\sqrt{N}}$). Thus,

22   if it is desired to reduce the statistical error by a factor of two, it will be necessary to increase the

23   number of samples N by a factor of four, which, in turn, increases the computational load that is

24   required to generate the pixel values, for each of the numerous pixels in the image.

1   Additionally, since the Monte Carlo method requires random numbers, an efficient

2 mechanism for generating random numbers is needed. Generally, digital computers are provided

3 with so-called "random number" generators, which are computer programs which can be processed

4 to generate a set of numbers that are approximately random. Since the random number generators

5 use deterministic techniques, the numbers that are generated are not truly random. However, the

6 property that subsequent random numbers from a random number generator are statistically

7 independent should be maintained by deterministic implementations of pseudo-random numbers on

8 a computer.

9   The Grabenstein application describes a computer graphics system and method for generating

10 pixel values for pixels in an image using a strictly deterministic methodology for generating sample

11 points, which avoids the above-described problems with the Monte Carlo method. The strictly

12 deterministic methodology described in the Grabenstein application provides a low-discrepancy

13 sample point sequence which ensures, *a priori*, that the sample points are generally more evenly

14 distributed throughout the region over which the respective integrals are being evaluated. In one

15 embodiment, the sample points that are used are based on a so-called Halton sequence. See, for

16 example, J. H. Halton, *Numerische Mathematik*, Vol. 2, pp. 84-90 (1960) and W. H. Press, et al.,

17 Numerical Recipes in Fortran (2d Edition) page 300 (Cambridge University Press, 1992). In a

18 Halton sequence generated for number base "p," where base "p" is a selected prime number, the "k-

19 th" value of the sequence, represented by H_p^k is generated by use of a "radical inverse" operation,

20 that is, by

21   (1)   writing the value "k" as a numerical representation of the value in the selected base

22   "p," thereby to provide a representation for the value as $D_M D_{M-1} \cdots D_2 D_1$, where

23   D_m (m=1, 2, ..., M) are the digits of the representation,

24   (2)   putting a radix point (corresponding to a decimal point for numbers written in base

25   ten) at the least significant end of the representation $D_M D_{M-1} \cdots D_2 D_1$ written in

26   step (1) above, and

1   (3)   reflecting the digits around the radix point to provide $0.D_1D_2\cdots D_{M-1}D_M$, which

2   corresponds to $H_p^{\,k}$.

3   It will be appreciated that, regardless of the base "p" selected for the representation, for any series

4   of values, one, two, ... "k," written in base "p," the least significant digits of the representation will

5   change at a faster rate than the most significant digits. As a result, in the Halton sequence $H_p^{\,1}$, $H_p^{\,2}$,

6   ...$H_p^{\,k}$, the most significant digits will change at the faster rate, so that the early values in the sequence

7   will be generally widely distributed over the interval from zero to one, and later values in the

8   sequence will fill in interstices among the earlier values in the sequence. Unlike the random or

9   pseudo-random numbers used in the Monte Carlo method as described above, the values of the

10   Halton sequence are not statistically independent; on the contrary, the values of the Halton sequence

11   are strictly deterministic, "maximally avoiding" each other over the interval, and so they will not

12   clump, whereas the random or pseudo-random numbers used in the Monte Carlo method may clump.

13   It will be appreciated that the Halton sequence as described above provides a sequence of

14   values over the interval from zero to one, inclusive along a single dimension. A multi-dimensional

15   Halton sequence can be generated in a similar manner, but using a different base for each dimension.

16   A generalized Halton sequence, of which the Halton sequence described above is a special

17   case, is generated as follows. For each starting point along the numerical interval from zero to one,

18   inclusive, a different Halton sequence is generated. Defining the pseudo-sum $x \oplus_p y$ for any x and y

19   over the interval from zero to one, inclusive, for any integer "p" having a value greater than two, the

20   pseudo-sum is formed by adding the digits representing "x" and "y" in reverse order, from the most-

21   significant digit to the least-significant digit, and for each addition also adding in the carry generated

22   from the sum of next more significant digits. Thus, if "x" in base "p" is represented by

23    $0.X_1X_2\cdots X_{M-1}X_M$, where each "$X_m$" is a digit in base "p," and if "y" in base "p" is represented

24   by $0.Y_1Y_2\cdots Y_{N-1}Y_N$, where each "Y_n" is a digit in base "p" (and where "M," the number of

25   digits in the representation of "x" in base "p", and "N," the number of digits in the representation of

26   "y" in base "p", may differ), then the pseudo-sum "z" is represented by $0.Z_1Z_2\cdots Z_{L-1}Z_L$,

27   where each "Z_l" is a digit in base "p" given by $Z_l = \left(X_l + Y_l + C_l \right) \bmod p$, where "mod"

1   represents the modulo function, and $C_l = \begin{cases} 1 \; for \; X_{l-1} + Y_{l-1} + Z_{l-1} \geq p \\ 0 \; otherwise \end{cases}$ is a carry value from the

2   "l-1st" digit position, with C_1 being set to zero.

3   Using the pseudo-sum function as described above, the generalized Halton sequence that is

4   used in the system described in the Grabenstein application is generated as follows. If "p" is an

5   integer, and x_0 is an arbitrary value on the interval from zero to one, inclusive, then the "p"-adic von

6   Neumann-Kakutani transformation $T_p(x)$ is given by

$$T_p(x) \quad := \quad x \oplus_p \frac{1}{p} \qquad\qquad (4),$$

8   and the generalized Halton sequence x_0, x_1, $x_{2,\ldots}$ is defined recursively as

$$x_{n+1} \quad = \quad T_p\big(x_n\big) \qquad\qquad (5)$$

10   From equations (4) and (5), it is clear that, for any value for "p," the generalized Halton sequence

11   can provide that a different sequence will be generated for each starting value of "x," that is, for each

12   x_0. It will be appreciated that the Halton sequence H_p^k as described above is a special case of the

13   generalized Halton sequence (equations (4) and (5)) for $x_0=0$.

14   The use of a strictly deterministic low-discrepancy sequence can provide a number of

15   advantages over the random or pseudo-random numbers that are used in connection with the Monte

16   Carlo technique. Unlike the random numbers used in connection with the Monte Carlo technique,

17   the low discrepancy sequences ensure that the sample points are more evenly distributed over a

18   respective region or time interval, thereby reducing error in the image which can result from

19   clumping of such sample points which can occur in the Monte Carlo technique. That can facilitate

20   the generation of images of improved quality when using the same number of sample points at the

21   same computational cost as in the Monte Carlo technique.

1 **SUMMARY OF THE INVENTION**

2   The invention provides a new and improved computer graphics system and method for
3 generating pixel values for pixels in the image using a strictly deterministic methodology for
4 generating sample points for use in evaluating integrals defining aspects of the image.

5   In brief summary, the invention provides, in one aspect, a computer graphics system for
6 generating a pixel value for a pixel in an image, the pixel being representative of a point in a scene
7 as recorded on an image plane of a simulated camera. The computer graphics system comprises a
8 sample point generator and a function evaluator. The sample point generator is configured to
9 generate a set of sample points representing at least one simulated element of the simulated camera,
10 the sample points representing elements of, illustratively, for sample points on the image plane,
11 during time interval during which the shutter is open, and on the lens, a Hammersley sequence, and,
12 for use in global illumination, a scrambled Halton sequence. The function evaluator configured to
13 generate at least one value representing an evaluation of said selected function at one of the sample
14 points generated by said sample point generator, the value generated by the function evaluator
15 corresponding to the pixel value.

16 **BRIEF DESCRIPTION OF THE DRAWINGS**

17   This invention is pointed out with particularity in the appended claims. The above and
18 further advantages of this invention may be better understood by referring to the following
19 description taken in conjunction with the accompanying drawings, in which:

20   FIG. 1 depicts an illustrative computer graphics system constructed in accordance with the
21 inventionn

22 **DETAILED DESCRIPTION OF AN ILLUSTRATIVE EMBODIMENT**

1   The invention provides an computer graphic system and method for generating pixel values

2   for pixels in an image of a scene, which makes use of a strictly-deterministic methodology for

3   generating sample points for use in generating sample values for evaluating the integral or integrals

4   whose function(s) represent the contributions of the light reflected from the various points in the

5   scene to the respective pixel value, rather than the random or pseudo-random Monte Carlo

6   methodology which has been used in the past. The strictly-deterministic methodology ensures *a*

7   *priori* that the sample points will be generally more evenly distributed over the interval or region

8   over which the integral(s) is (are) to be evaluated in a low-discrepancy manner.

9   FIG. 1 attached hereto depicts an illustrative computer system 10 that makes use of such a

10   strictly deterministic methodology. With reference to FIG. 1, the computer system 10 in one

11   embodiment includes a processor module 11 and operator interface elements comprising operator

12   input components such as a keyboard 12A and/or a mouse 12B (generally identified as operator input

13   element(s) 12) and an operator output element such as a video display device 13. The illustrative

14   computer system 10 is of the conventional stored-program computer architecture. The processor

15   module 11 includes, for example, one or more processor, memory and mass storage devices, such

16   as disk and/or tape storage elements (not separately shown), which perform processing and storage

17   operations in connection with digital data provided thereto. The operator input element(s) 12 are

18   provided to permit an operator to input information for processing. The video display device 13 is

19   provided to display output information generated by the processor module 11 on a screen 14 to the

20   operator, including data that the operator may input for processing, information that the operator may

21   input to control processing, as well as information generated during processing. The processor

22   module 11 generates information for display by the video display device 13 using a so-called

23   "graphical user interface" ("GUI"), in which information for various applications programs is

24   displayed using various "windows." Although the computer system 10 is shown as comprising

25   particular components, such as the keyboard 12A and mouse 12B for receiving input information

26   from an operator, and a video display device 13 for displaying output information to the operator,

27   it will be appreciated that the computer system 10 may include a variety of components in addition

28   to or instead of those depicted in FIG. 1.

1   In addition, the processor module 11 includes one or more network ports, generally identified

2   by reference numeral 14, which are connected to communication links which connect the computer

3   system 10 in a computer network. The network ports enable the computer system 10 to transmit

4   information to, and receive information from, other computer systems and other devices in the

5   network. In a typical network organized according to, for example, the client-server paradigm,

6   certain computer systems in the network are designated as servers, which store data and programs

7   (generally, "information") for processing by the other, client computer systems, thereby to enable

8   the client computer systems to conveniently share the information. A client computer system which

9   needs access to information maintained by a particular server will enable the server to download the

10   information to it over the network. After processing the data, the client computer system may also

11   return the processed data to the server for storage. In addition to computer systems (including the

12   above-described servers and clients), a network may also include, for example, printers and facsimile

13   devices, digital audio or video storage and distribution devices, and the like, which may be shared

14   among the various computer systems connected in the network. The communication links

15   interconnecting the computer systems in the network may, as is conventional, comprise any

16   convenient information-carrying medium, including wires, optical fibers or other media for carrying

17   signals among the computer systems. Computer systems transfer information over the network by

18   means of messages transferred over the communication links, with each message including

19   information and an identifier identifying the device to receive the message.

20   It will be helpful to initially provide some background on operations performed by the

21   computer graphics system in generating an image. Generally, the computer graphic system

22   generates an image attempts to simulate an image of a scene that would be generated by a camera.

23   The camera includes a shutter that will be open for a predetermined time T starting at a time t_0 to

24   allow light from the scene to be directed to an image plane. The camera may also include a lens that

25   serves to focus light from the scene onto the image plane. The average radiance flux $L_{m,n}$ through

26   a pixel at position (m,n) on an image plane P, which represents the plane of the camera's recording

27   medium, is determined by

28   $$L_{m,n} = \frac{1}{|A_P| \cdot T \cdot |A_L|} \int_{A_P} \int_{t_0}^{t_0+T} \int_{A_L} L\big(h(x,t,y) - \omega(x,t,y)\big) f_{m,n}(x)\, dy\, dt\, dx \quad (6)$$

-9-

1   where "A_P" refers to the area of the pixel, A_L refers to the area of the portion of the lens through

2   which rays of light pass from the scene to the pixel, and $f_{m,n}$ represents a filtering kernel associated

3   with the pixel. An examination of the integral in equation (6) will reveal that the variables of

4   integration, "x," "y" and "t," are essentially dummy variables with the variable "y" referring to

5   integration over the lens area (A_L), the variable "t" referring to integration over time (the time

6   interval from t_0 to t_0+T) and the variable "x" referring to integration over the pixel area (A_P).

7   The value of the integral in equation (6) is approximated by identifying N_P sample points x_i

8   in the pixel area, and, for each sample point, shooting N_T rays at times $t_{i,j}$ in the time interval t_0 to

9   t_0+T through the focus into the scene, with each ray spanning N_L sample points $y_{i,j,k}$ on the lens area

10   A_L. The manner in which subpixel jitter positions x_i, points in time $t_{i,j}$ and positions on the lens $y_{i,j,k}$

11   are determined will be described below. These three parameters determine the primary ray hitting

12   the scene geometry in $h(x_i,t_{i,j},y_{i,j,k})$ with the ray direction $\omega(x_i,t_{i,j},y_{i,j,k})$. In this manner, the value of

13   the integral in equation (6) can be approximated as

$$L_{m,n} \approx \frac{1}{N} \sum_{i=0}^{N_P-1} \frac{1}{N_T} \sum_{j=0}^{N_T-1} \frac{1}{N_L} \sum_{k=0}^{N_L-1} L\left(h\left(x_i,t_{i,j},y_{i,j,k}\right),-\omega\left(x_i,t_{i,j},y_{i,j,k}\right)\right) f_{m,n}\left(x_i\right) \quad (7),$$

15   where "N" is the total number of rays directed at the pixel.

16   It will be appreciated that rays directed from the scene toward the image can comprise rays

17   directly from one or more light sources in the scene, as well as rays reflected off surfaces of objects

18   in the scene. In addition, it will be appreciated that a ray that is reflected off a surface may have

19   been directed to the surface directly from a light source, or a ray that was reflected off another

20   surface. For a surface that reflects light rays, a reflection operator T_{fr} is defined that includes a

21   diffuse portion T_{fd}, a glossy portion T_{fg} and a specular portion T_{fs}, or

$$T_{f_r} = T_{f_d} + T_{f_g} + T_{f_s} \qquad (8).$$

23   In that case, the Fredholm integral equation $L=L_e+T_{fr}L$ governing light transport can be represented

24   as

$$L = L_e + T_{f_r-f_s}L_e + T_{f_s}\left(L - L_e\right) + T_{f_s}L + T_{f_d}T_{f_g+f_s}L + T_{f_d}T_{f_d}L \tag{9},$$

where transparency has been ignored for the sake of simplicity; transparency is treated in an analogous manner. The individual terms in equation (9) are

  (i) L_e represents flux due to a light source;

  (ii) $T_{f_r-f_s}L_e$ (where $T_{f_r-f_s} = T_{f_r} - T_{f_s}$) represents direct illumination, that is, flux reflected off a surface that was provided thereto directly by a light source; the specular component, associated with the specular portion T_{fs} of the reflection operator, will be treated separately since it is modeled as a δ-distribution;

  (iii) $T_{f_g}\left(L - L_e\right)$ represents glossy illumination, which is handled by recursive distribution ray tracing, where, in the recursion, the source illumination has already been accounted for by the direct illumination (item (ii) above);

  (iv) $T_{f_s}L$ represents a specular component, which is handled by recursively using "L" for the reflected ray;

  (v) $T_{f_d}T_{f_g+f_s}L$ (where $T_{f_g+f_s} = T_{f_g} + T_{f_s}$) represents a caustic component, which is a ray that has been reflected off a glossy or specular surface (reference the $T_{f_g+f_s}$ operator) before hitting a diffuse surface (reference the T_{f_d} operator); this contribution can be approximated by a high resolution caustic photon map; and

  (vi) $T_{f_d}T_{f_d}L$ represents ambient light, which is very smooth and is therefore approximated using a low resolution global photon map.

1   As noted above, the value integral (equation (6)) is approximated by solving equation (7)

2   making use of sample points x_i, $t_{i,j}$ and $y_{i,j,k}$, where "x_i" refers to sample points within area A_L of the

3   respective pixel at location (m,n) in the image plane, "$t_{i,j}$" refers to sample points within the time

4   interval t_0 to t_0+T during which the shutter is open, and "$y_{i,j,k}$" refers to sample points on the lens A_L.

5   In accordance with one aspect of the invention, the sample points x_i comprise two-dimensional

6   Hammersley points, which are defined as $\left(\dfrac{i}{N}, \Phi_2(i) \right)$, where $0 \leq i < N := \left(2^n \right)^2$ for a selected

7   integer parameter "n," and $\Phi_2(i)$ refers to the radical inverse of "i" in base "two." Generally, the "s"

8   dimensional Hammersley point set is defined as

$$U_{N,s}^{Hammersley} : \{0, \cdots, N-1\} \quad \rightarrow \quad I^s$$

$$i \quad \mapsto \quad x_i := \left(\frac{i}{N}, \Phi_{b_1}(i), \cdots, \Phi_{b_{s-1}}(i) \right) \qquad (10),$$

10   where I^s is the s-dimensional unit cube [0,1) (that is, an s-dimensional cube that includes "zero," and

11   excludes "one"), the number of points "N" in the set is fixed and $b_1, ..., b_{s-1}$ are bases. The bases do

12   not need to be prime numbers, but they are preferably relatively prime to provide a relatively

13   uniform distribution. The radical inverse function Φ_b, in turn, is generally defined as

14

$$\Phi_b : N_0 \quad \rightarrow \quad I$$

$$i = \sum_{j=0}^{\infty} a_j(i)\, b^j \quad \mapsto \quad \sum_{j=0}^{\infty} a_j(i)\, b^{-j-1} \qquad (11)$$

15   where $\left(a_j \right)_{j=0}^{\infty}$ is the representation of "i" in integer base "b." The two-dimensional Hammersley

16   points form a stratified pattern on a 2^n by 2^n grid. Considering the grid as subpixels, the complete

17   subpixel grid underlying the image plane can be filled by simply abutting copies of the grid to each

18   other.

1   Given integer subpixel coordinates (s_x, s_y) the instance "i" and coordinates (x,y) for the sample

2   point x_i in the image plane can be determined as follows. Preliminarily, examining

3        $\left(\dfrac{i}{N}, \Phi_2(i) \right)_{i=0}^{N-1}$, one observes that

4   (a) each line in the stratified pattern is a shifted copy of another, and

5   (b) the pattern is symmetric to the line y=x, that is, each column is a shifted copy of another

6   column.

7   Accordingly, given the integer permutation $\sigma(k) := 2^n \, \Phi_2(k)$ for $0 \le k < 2^n$, subpixel coordinates

8   (s_x, s_y) are mapped onto strata coordinates $(j,k) := \left(s_x \bmod 2^n, s_y \bmod 2^n \right)$, an instance number

9   "i" is computed as

10
$$i = j \, 2^n + \sigma(k) \tag{12}$$

11   and the positions of the jittered subpixel sample points are determined according to

12
$$x_i = \left(s_x + \Phi_2(k), s_y + \Phi_2(j) \right)$$
$$= \left(s_x + \frac{\sigma(k)}{2^n}, s_y + \frac{\sigma(j)}{2^n} \right) \tag{13}.$$

13   An efficient algorithm for generating the positions of the jittered subpixel sample points x_i will be

14   provided below in connection with Code Segment 1. A pattern of sample points whose positions

15   are determined as described above in connection with equations (12) and (13) has an advantage of

16   having much reduced discrepancy over a pattern determined using a Halton sequence or windowed

17   Halton sequence, as described in the aforementioned Grabenstein application, and therefore the

1   approximation described above in connection with equation (7) gives in general a better estimation

2   to the value of the integral described above in connection with equation (6). In addition, if "n" is

3   sufficiently large, sample points in adjacent pixels will have different patterns, reducing the

4   likelihood that undesirable artifacts will be generated in the image.

5   A ray tree is a collection of paths of light rays that are traced from a point on the simulated

6   camera's image plane into the scene. The computer graphics system 10 generates a ray tree by

7   recursively following transmission, subsequent reflection and shadow rays using trajectory splitting.

8   In accordance with another aspect of the invention, a path is determined by the components of one

9   vector of a global generalized scrambled Hammersley point set. Generally, a scrambled

10 Hammersley point set reduces or eliminates a problem that can arise in connection with higher-

11 dimensioned low-discrepancy sequences since the radical inverse function Φ_b typically has

12 subsequences of b-1 equidistant values spaced by $\frac{1}{b}$. Although these correlation patterns are

13 merely noticeable in the full s-dimensional space, they are undesirable since they are prone to

14 aliasing. The computer graphics system 10 attenuates this effect by scrambling, which corresponds

15 to application of a permutation to the digits of the b-ary representation used in the radical inversion.

16 For a permutation σ from a symmetric group S_b over integers 0,...,b-1, the scrambled radical inverse

17 is defined as

$$\Phi_b : N_0 \times S_b \quad \rightarrow \quad I$$

$$(i,\sigma) \quad \mapsto \quad \sum_{j=0}^{\infty} \sigma\left(a_j(i)\right) b^{-j-1} \quad \Leftrightarrow \quad i = \sum_{j=0}^{\infty} a_j(i)\, b^j \qquad \textbf{(14)}.$$

18 If the permutation "σ" is the identity, the scrambled radical inverse corresponds to the unscrambled

19 radical inverse. In one embodiment, the computer graphics system generates the permutation σ

20 recursively as follows. Starting from the permutation $\sigma_2 = (0,2)$ for base b=2, the sequence of

21 permutations is defined as follows:

1   (i)   if the base "b" is even, the permutation σ_b is generated by first taking the values of $2\,\sigma_{\frac{b}{2}}$

2   and appending the values of $2\,\sigma_{\frac{b}{2}} + 1$, and

3   (ii)   if the base "b" is odd, the permutation σ_b is generated by taking the values of σ_{b-1},

4   incrementing each value that is greater than or equal to $\dfrac{b-1}{2}$ by one, and inserting

5   the value b-1 in the middle.

6   This recursive procedure results in

$\sigma_2 = (0,1)$

$\sigma_3 = (0,1,2)$

$\sigma_4 = (0,2,1,3)$

$\sigma_5 = (0,3,2,1,4)$

$\sigma_6 = (0,2,4,1,3,5)$

$\sigma_7 = (0,2,5,3,1,4,6)$

$\sigma_8 = (0,4,2,6,1,5,3,7)....$

14   The computer graphics system 10 can generate a generalized low-discrepancy point set as

15   follows. It is possible to obtain a low-discrepancy sequence by taking any rational s-dimensional

16   point "x" as a starting point and determine a successor by applying the corresponding incremental

17   radical inverse function to the components of "x." The result is referred to as the generalized low-

18   discrepancy point set. This can be applied to both the Halton sequence and the Hammersley

19   sequence. In the case of the generalized Halton sequence, this can be formalized as

20

$$x_i \;=\; \left(\Phi_{b_1}\left(i + i_1\right), \Phi_{b_2}\left(i + i_2\right), \ldots, \Phi\left(i + i_s\right) \right) \qquad (15),$$

1   where the integer vector $(i_1, i_2, ..., i_s)$ represents the offsets per component and is fixed in advance

2   for a generalized sequence. The integer vector can be determined by applying the inverse of the

3   radical inversion to the starting point "x." A generalized Hammersley sequence can be generated

4   in an analogous manner.

5   Returning to trajectory splitting, generally trajectory splitting is the evaluation of a local

6   integral, which is of small dimension and which makes the actual integrand smoother, which

7   improves overall convergence. Applying replication, positions of low-discrepancy sample points

8   are determined that can be used in evaluating the local integral. The low-discrepancy sample points

9   are shifted by the corresponding elements of the global scrambled Hammersley point set. Since

10   trajectory splitting can occur multiple times on the same level in the ray tree, branches of the ray tree

11   are decorrelated in order to avoid artifacts, the decorrelation being accomplished by generalizing the

12   global scrambled Hammersley point set.

13   An efficient algorithm for generating a ray tree will be provided below in connection with

14   Code Segment 2. Generally, in that algorithm, the instance number "i" of the low-discrepancy

15   vector, as determined above in connection with equation (12), and the number "d" of used

16   components, which corresponds to the current integral dimension, are added to the data structure that

17   is maintained for the respective ray in the ray tree. The ray tree of a subpixel sample is completely

18   specified by the instance number "i." After the dimension has been set to "two," which determines

19   the component of the global Hammersley point set that is to be used next, the primary ray is cast into

20   the scene to span its ray tree. In determining the deterministic splitting by the components of low

21   discrepancy sample points, the computer graphics system 10 initially allocates the required number

22   of dimensions "Δd." For example, in simulating glossy scattering, the required number of

23   dimensions will correspond to "two." Thereafter, the computer graphics system 10 generates

24   scattering directions from the offset given by the scrambled radical inverses

25   $\Phi_{b_d}\left(i, \sigma_{b_d}\right), ..., \Phi_{b_{d+\Delta d-1}}\left(i, \sigma_{b_{d+\Delta d-1}}\right)$, yielding the instances

26
$$\left(y_{i,j}\right)_{j=0}^{M-1} = \left(\Phi_{b_d}\left(i, \sigma_{b_d}\right) \oplus \frac{j}{M}, ..., \Phi_{b_{d+\Delta d-1}}\left(i, \sigma_{b_{d+\Delta d-1}}\right) \oplus \Phi_{b_{d+\Delta d-2}}\left(i, \sigma_{b_{d+\Delta d-2}}\right)\right) \quad (16),$$

1 where "⊕" refers to addition modulo "one." Each direction of the "M" replicated rays is determined

2 by $y_{i,j}$ and enters the next level of the ray tree with $d' := d + \Delta d$ as the new integral dimension in

3 order to use the next elements of the low-discrepancy vector, and i' = i + j in order to decorrelate

4 subsequent trajectories. Using an infinite sequence of low-discrepancy sample points, the replication

5 heuristic is turned into an adaptive consistent sampling arrangement. That is, computer graphics

6 system 10 can fix the sampling rate ΔM, compare current and previous estimates every ΔM samples,

7 and, if the estimates differ by less than a predetermined threshold value, terminate sampling. The

8 computer graphics system 10 can, in turn, determine the threshold value, by importance information,

9 that is, how much the local integral contributes to the global integral.

10 As noted above, the integral described above in connection with equation (6) is over a finite

11 time period T from t_0 to t_0+T, during which time the shutter of the simulated camera is open. During

12 the time period, if an object in the scene moves, the moving object may preferably be depicted in the

13 image as blurred, with the extent of blurring being a function of the object's motion and the time

14 interval t_0+T. Generally, motion during the time an image is recorded is linearly approximated by

15 motion vectors, in which case the integrand (equation (6)) is relatively smooth over the time the

16 shutter is open and is suited for correlated sampling. For a ray instance "i," started at the subpixel

17 position x_i, the offset $\Phi_3(i)$ into the time interval is generated and the N_T-1 subsequent samples

18 $\Phi_3(i) + \dfrac{j}{N_T} \bmod 1$ for $0 \leq j < N_T$ that is

19
$$t_{i,j} := t_0 + \left(\Phi_3(i) \oplus \frac{j}{N_T} \right) \cdot T \tag{17}.$$

20 Determining sample points in this manner fills the sampling space, resulting in a more rapid

21 convergence to the value of the integral (equation (6)). For subsequent trajectory splitting, rays are

22 decorrelated by setting the instance i' = i + j.

23 In addition to determining the position of the jittered subpixel sample point x_i, and adjusting

24 the camera and scene according to the sample point $t_{i,j}$ for the time, the computer graphics system

1   also simulates depth of field. In simulating depth of field, the camera to be simulated is assumed

2   to be provided with a lens having known optical characteristics and, using geometrical optics, the

3   subpixel sample point x_i is mapped through the lens to yield a mapped point x_i'. The lens is sampled

4   by mapping the dependent samples

5

$$y_{i,j,k} = \left(\left(\Phi_5(i+j, \sigma_5) \oplus \frac{k}{N_L} \right), \left(\Phi_7(i+j, \sigma_7) \oplus \Phi_2(k) \right) \right) \tag{18}$$

6   onto the lens area A_L using a suitable one of a plurality of known transformations. Thereafter, a ray

7   is shot from the sample point on the lens specified by $y_{i,j,k}$ through the point x_i' into the scene. The

8   offset $\left(\Phi_5(i+j, \sigma_5), \Phi_7(i+j, \sigma_7) \right)$ in equation (18) comprise the next components taken from

9   the generalized scrambled Hammersley point set, which, for trajectory splitting, is displaced by the

10   elements $\left(\frac{k}{N_L}, \Phi_2(k) \right)$ of the two-dimensional Hammersley point set. The instance of the ray

11   originating from sample point $y_{i,j,k}$ is set to "i+j+k" in order to decorrelate further splitting down the

12   ray tree. In equation (18), the scrambled samples $\left(\Phi_5(i+j, \sigma_5), \Phi_7(i+j, \sigma_7) \right)$ are used instead

13   of the unscrambled samples of $\left(\Phi_5(i+j), \Phi_7(i+j) \right)$ since in bases "five" and "seven" the up to

14   five unscrambled samples will lie on a straight line, which will not be the case for the scrambled

15   samples.

16   In connection with determination of a value for the direct illumination ($T_{f_r-f_s} L_e$ above),

17   direct illumination is represented as an integral over the surface of the scene ∂V , which integral is

18   decomposed into a sum of integrals, each over one of the "L" single area light sources in the scene.

19   The individual integrals in turn are evaluated by dependent sampling, that is

$$\left(T_{f_r - f_s} L_e\right)(y,z) = \int_{\partial V} L_e(x,y) f_r(x,y,z) G(x,y) \, dx$$

$$= \sum_{k=1}^{L} \int_{\text{supp} L_{e,k}} L_e(x,y) f_r(x,y,z) G(x,y) \, dx \qquad \textbf{(19)}.$$

$$\approx \sum_{k=1}^{L} \frac{1}{M_K} \sum_{j=0}^{M_k - 1} L_e(x_j,y) f_r(x_j,y,z) G(x_j,y)$$

2   where $\text{supp}L_{e,k}$ refers to the surface of the respective "k-th" light source. In evaluating the estimate

3   of the integral for the "k-th" light source, for the M_k-th query ray, shadow rays determine the fraction

4   of visibility of the area light source, since the pont visibility varies much more than the smooth

5   shadow effect. For each light source, the emission L_e is attenuated by a geometric term G, which

6   includes the visibility, and the surface properties are given by a bidirectional distribution function

7   f_r-f_s. These integrals are local integrals in the ray tree yielding the value of one node in the ray tree,

8   and can be efficiently evaluated using dependent sampling. In dependent sampling, the query ray

9   comes with the next free integral dimension "d" and the instance "i," from which the dependent

10   samples are determined in accordance with

$$x_j = \left(\Phi_{b_d}\left(i,\sigma_{b_d}\right) \oplus \frac{j}{M_k}, \Phi_{b_{d+1}}\left(i,\sigma_{b_{d+1}}\right) \oplus \Phi_2(j) \right) \qquad \textbf{(20)}.$$

12   The offset $\left(\Phi_{b_d}\left(i,\sigma_{b_d}\right), \Phi_{b_{d+1}}\left(i,\sigma_{b_{d+1}}\right) \right)$ again is taken from the corresponding generalized

13   scrambled Hammersley point set, which shifts the two-dimensional Hammersley point set

14    $\left(\frac{j}{M_k}, \Phi_2(j) \right)$ on the light source. Selecting the sample rate $M_k = 2^{n_k}$ as a power of two, the

15   local minima is obtained for the discrepancy of the Hammersley point set that perfectly stratifies the

16   light source. As an alternative, the light source can be sampled using an arbitrarily-chosen number

1   M_k of sample points using $\left(\dfrac{j}{M_k}, \Phi_{M_K}\left(j, \sigma_{M_k}\right) \right)_{j=0}^{M_k - 1}$ as a replication rule. Due to the implicit

2   stratification of the positions of the sample points as described above, the local convergence will be

3   very smooth.

4   The glossy contribution $T_{f_e}\left(L - L_e\right)$ is determined in a manner similar to that described

5   above in connection with area light sources (equations (19) and (20), except that a model f_g used to

6   simulate glossy scattering will be used instead of the bidirectional distribution function f_r. In

7   determining the glossy contribution, two-dimensional Hammersley points are generated for a fixed

8   splitting rate M and shifted modulo "one" by the offset $\left(\Phi_{b_d}\left(i, \sigma_{b_d}\right), \Phi_{b_{d+1}}\left(i, \sigma_{b_{d+1}}\right) \right)$ taken from

9   the current ray tree depth given by the dimension field "d" of the incoming ray. The ray trees

10   spanned into the scattering direction are decorrelated by assigning the instance fields i'=i+j in a

11   manner similar to that done for simulation of motion blur and depth of field, as described above.

12   The estimates generated for all rays are averaged by the splitting rate "M" and propagated up the ray

13   tree.

14   Volumetric effects are typically provided by performing a line integration along respective

15   rays from their origins to the nearest surface point in the scene. In providing for a volumetric effect,

16   the computer graphics system 10 generates from the ray data a corresponding offset $\Phi_{b_d}\left(i\right)$ which

17   it then uses to shift the M equidistant samples on the unit interval seen as a one-dimensional torus.

18   In doing so, the rendering time is reduced in comparison to use of an uncorrelated jittering

19   methodology.

20   Global illumination includes a class of optical effects, such as indirect illumination, diffuse

21   and glossy inter-reflections, caustics and color bleeding, that the computer graphics system 10

22   simulates in generating an image of objects in a scene. Simulation of global illumination typically

1   involves the evaluation of a rendering equation. For the general form of an illustrative rendering

2   equation useful in global illumination simulation, namely:

3 $$L(\vec{x},\vec{w}) \;=\; L_e(\vec{x},\vec{w}) + \int_{S'} f(\vec{x},\vec{w}' \to \vec{w})\, G(\vec{x},\vec{x}')\, V(\vec{x},\vec{x}')\, L(\vec{x},\vec{w}')\, dA' \qquad \text{(21)}$$

4   it is recognized that the light radiated at a particular point $\vec{x}$ in a scene is generally the sum of two

5   components, namely, the amount of light (if any) that is emitted from the point and the amount of

6   light (if any) that originates from all other points and which is reflected or otherwise scattered from

7   the point $\vec{x}$. In equation (21), $L(\vec{x},\vec{w})$ represents the radiance at the point $\vec{x}$ in the direction

8     $\vec{w} = (\theta,\varphi)$ (where "θ" represents the angle of direction $\vec{w}$ relative to a direction orthogonal of the

9   surface of the object in the scene containing the point $\vec{x}$, and "φ" represents the angle of the

10   component of direction $\vec{w}$ in a plane tangential to the point $\vec{x}$). Similarly, $L(\vec{x}',\vec{w}')$ in the

11   integral represents the radiance at the point $\vec{x}'$ in the direction $\vec{w}' = (\theta',\varphi')$ (where "θ'" represents

12   the angle of direction $\vec{w}'$ relative to a direction orthogonal of the surface of the object in the scene

13   containing the point $\vec{x}'$, and "φ'" represents the angle of the component of direction $\vec{w}'$ in a plane

14   tangential to the point $\vec{x}'$), and represents the light, if any, that is emitted from point $\vec{x}'$ which may

15   be reflected or otherwise scattered from point $\vec{x}$.

16   Continuing with equation (21), $L_e(\vec{x},\vec{w})$ represents the first component of the sum, namely,

17   the radiance due to emission from the point $\vec{x}$ in the direction $\vec{w}$, and the integral over the sphere

18   S' represents the second component, namely, the radiance due to scattering of light at point $\vec{x}$.

19     $f(\vec{x},\vec{w}' \to \vec{w})$ is a bidirectional scattering distribution function which describes how much of the

20   light coming from direction $\vec{w}'$ is reflected, refracted or otherwise scattered in the direction $\vec{w}$, and

1   is generally the sum of a diffuse component, a glossy component and a specular component. In

2   equation (21), the function $G(\vec{x}, \vec{x}')$ is a geometric term

3

$$G(\vec{x}, \vec{x}') \;=\; \frac{\cos\theta \; \cos\theta'}{|\vec{x} - \vec{x}'|^2} \tag{22}$$

4   where θ and θ' are angles relative to the normals of the respective surfaces at points $\vec{x}$ and $\vec{x}'$,

5   respectively. Further in equation (21), $V(\vec{x}, \vec{x}')$ is a visibility function which equals the value one

6   if the point $\vec{x}'$ is visible from the point $\vec{x}$ and zero if the point $\vec{x}'$ is not visible from the point $\vec{x}$.

7   The computer graphics system 10 makes use of global illumination specifically in connection

8   with determination of the diffuse component $T_{f_d} T_{f_d} L$ and the caustic component $T_{f_d} T_{f_g + f_s} L$

9   using a photon map technique. Generally, a photon map is constructed by simulating the emission

10   of photons by light source(s) in the scene and tracing the path of each of the photons. For each

11   simulated photon that strikes a surface of an object in the scene, information concerning the

12   simulated photon is stored in a data structure referred to as a photon map, including, for example,

13   the simulated photon's color, position and direction angle. Thereafter a Russian roulette operation

14   is performed to determine the photon's state, that is, whether the simulated photon will be deemed

15   to have been absorbed or reflected by the surface. If a simulated photon is deemed to have been

16   reflected by the surface, the simulated photon's direction is determined using, for example, a

17   bidirectional reflectance distribution function ("BRDF"). If the reflected simulated photon strikes

18   another surface, these operations will be repeated (reference the aforementioned Grabenstein

19   application). The data structure in which information for the various simulated photons is stored

20   may have any convenient form; typically k-dimensional trees, for "k" an integer" are used. After the

21   photon map has been generated, it can be used in rendering the respective components of the image.

22   In generating a photon map, the computer graphics system 10 simulates photons trajectories,

23   thus avoiding the necessity of discretizing the kernel of the underlying integral equation. The

24   interactions of the photons with the scene, as described above, are stored and used for density

1   estimation. The computer graphics system 10 makes use of a scrambled Halton sequence, which has

2   better discrepancy properties in higher dimensions than does an unscrambled Halton sequence. The

3   scrambled Halton sequence also has the benefit, over a random sequence, that the approximation

4   error decreases more smoothly, which will allow for use of an adaptive termination scheme during

5   generation of the estimate of the integral. In addition, since the scrambled Halton sequence is strictly

6   deterministic, generation of estimates can readily be parallelized by assigning certain segments of

7   the low-discrepancy sequence to ones of a plurality of processors, which can operate on portions of

8   the computation independently and in parallel. Since usually the space in which photons will be shot

9   by selecting directions will be much larger than the area of the light sources from which the photons

10   were initially shot, it is advantageous to make use of components of smaller discrepancy, for

11   example, Φ_2 or Φ_3 (where, as above, Φ_b refers to the radical inverse function for base "b"), for

12   angular scattering and components of higher discrepancy, for example, Φ_5 or Φ_7 for area sampling,

13   which will facilitate filling the space more uniformly.

     The computer graphics system 10 estimates the radiance from the photons in accordance with

$$\overline{L}_r(x,\omega) \approx \frac{1}{A} \sum_{i \in B_k(x)} f_r(\omega_i, x, \omega) \Phi_i \tag{23}$$

16   where, in equation (23), Φ_i represents the energy of the respective "i-th" photon, ω_i is the direction

17   of incidence of the "i-th photon, "$B_k(x)$ represents the set of the "k" nearest photons around the point

18   "x," and "A" represents an area around point "x" that includes the photons in the set $B_k(x)$. The

19   computer graphics system 10 makes use of an unbiased but consistent estimator is used for **, which

20   is determined as follows. Taking the radius $r(B_k(x))$ of the query ball (**what is this?**) a tangential

21   disk D of radius $r(B_k(x))$ centered on the point x is divided into M equal-sized subdomains D_i, that

22   is

23

$$\cup_{i=0}^{M-1} D_i = D \quad and \quad D_i \cap D_j \neq 0 \; for \; i \neq j, \quad where \; |D_i| = \frac{|D|}{M} = \frac{\pi \, r^2(B_k(x))}{M} \tag{24}.$$

24   The set

1

$$P = \left\{ D_i \ \middle| \ D_i \cap \left\{ x_i \middle|_D \ \middle| \ i \in B_k(x) \right\} \neq 0 \right\} \qquad (25)$$

2 contains all the subdomains D_i that contain a point $x_i|_D$ on the disk, which is the position of the "i-th"

3 photon projected onto the plane defined by the disk D along its direction of incidence ω_i. Preferably,

4 the number M of subdomains will be on the order of $\sqrt{k}$ and the angular subdivision will be finer

5 than the radial subdivision in order to capture geometry borders. The actual area A is then

6 determined by

7

$$A = \pi r^2 (B_k(x)) \frac{|P|}{M} \qquad (26).$$

8 Determining the actual coverage of the disk D by photons significantly improves the radiance

9 estimate (equation (23)) in corners and on borders, where the area obtained by the standard estimate

10 $\pi r^2 (B_k(x))$ would be too small, which would be the case at corners, or too large, which would

11 be the case at borders. In order to avoid blurring of sharp contours of caustics and shadows, the

12 computer graphics system 10 sets the radiance estimate L to black if all domains D_i that touch x do

13 not contain any photons.

14 It will be appreciated that, in regions of high photon density, the "k" nearest photons may

15 lead to a radius $r(B_k(x)$ that is nearly zero, which can cause an over-modulation of the estimate.

16 Over-modulation can be avoided by selecting a minimum radius r_{min}, which will be used if $r(B_k(x))$

17 is less than r_{min}. In that case, instead of equation (23), the estimate is generated in accordance with

18

$$\overline{L}_r(x, \omega) = \frac{N}{k} \sum_{i \in B_k(x)} \Phi_i \ f_r(\omega_i, x, \omega_r) \qquad (27)$$

19 assuming each photon is started with $\frac{1}{N}$ of the total flux Φ. The estimator in equation (27)

20 provides an estimate for the mean flux of the "k" photons if $r(B_k(x)) < r_{min}$.

1   The global photon map is generally rather coarse and, as a result, subpixel samples can result

2   in identical photon map queries. As a result, the direct visualization of the global photon map is

3   blurry and it is advantageous to perform a smoothing operation in connection therewith In

4   performing such an operation, the computer graphics system 10 performs a local pass integration that

5   removes artifacts of the direct visualization. Accordingly, the computer graphics system 10

6   generates an approximation for the diffuse illumination term $T_{f_d} T_{f_d} L$ as

7

$$T_{f_d} T_{f_d} L \approx \left(T_{f_d} \overline{L}_r \right)(x) = \int_{S^2(x)} f_d(x) \, \overline{L}_r\left(h(x,\vec{\omega}) \right) \cos\theta \, d\omega$$

$$\approx \frac{f_d(x)}{M} \sum_{i=0}^{M-1} \overline{L}_r\left(h\left(x, \omega\left(\arcsin \sqrt{u_{i,1}}, 2\pi u_{i,2} \right) \right) \right) \tag{28},$$

with the integral over the hemisphere $S^2(x)$ of incident directions aligned by the surface normal in

"x" being evaluated using importance sampling. The computer graphics system 10 stratifies the

sample points on a two-dimensional grid by applying dependent trajectory splitting with the

Hammersley sequence and thereafter applies irradiance interpolation. Instead of storing the incident

flux Φ_i of the respective photons, the computer graphics system 10 stores their reflected diffuse

power $f_d(x_i)\Phi_i$ with the respective photons in the photon map, which allows for a more exact

approximation than can be obtained by only sampling the diffuse BRDF in the hit points of the final

15   gather rays. In addition, the BRDF evaluation is needed only once per photon, saving the

16   evaluations during the final gathering. Instead of sampling the full grid, the computer graphics

17   system 10 uses adaptive sampling, in which refinement is triggered by contrast, distance traveled by

18   the final gather rays in order to more evenly sample the projected solid angle, and the number of

19   photons that are incident form the portion of the projected hemisphere. The computer graphics

20   system fills in positions in the grid that are not sampled by interpolation. The resulting image matrix

21   of the projected hemisphere is median filtered in order to remove weak singularities, after which the

22   approximation is generated . The computer graphics system 10 performs the same operation in

23   connection with hemispherical sky illumination.

24   The computer graphics system 10 processes final gather rays that strike objects that do not

25   cause caustics, such as plane glass windows, by recursive ray tracing. If the hit point of a final

1   gather ray is closer to its origin than a predetermined threshold, the computer graphics system 10

2   also performs recursive ray tracing. This reduces the likelihood that blurry artifacts will appear in

3   corners, which might otherwise occur since for close hit points the same photons would be collected,

4   which can indirectly expose the blurry structure of the global photon map.

5   Generally, photon maps have been taken as a snapshot at one point in time, and thus were

6   unsuitable in connection with rendering of motion blur. Following the observation that averaging

7   the result of a plurality of photon maps is generally similar to querying one photon map with the

8   total number of photons from all of the plurality of photon maps, the computer graphics system 10

9   generates N_T photon maps, where N_T is determined as described above, at points in time

$$t_b = t_0 + \frac{b + \frac{1}{2}}{N_T} T \qquad (29),$$

11   for $0 \leq b < N_T$. During rendering, the computer graphics system 10 uses the photon map with the

12   smallest time difference $\left| t_{i,J} - t_b \right|$ in connection with rendering for the time sample point $t_{i,j}$.

13   The invention provides a number of advantages. In particular, the invention provides a

14   computer graphics system that makes use of strictly deterministic distributed ray tracing based on

15   low-discrepancy sampling and dependent trajectory splitting in connection with rendering of an

16   image of a scene. Generally, strictly deterministic distributed ray tracing based on low-discrepancy

17   sampling and dependent trajectory splitting is simpler to implement than an implementation based

18   on random or pseudo-random numbers. Due to the properties of the radical inversion function,

19   stratification of sample points is intrinsic and does not need to be considered independently of the

20   generation of the positions of the sample points. In addition, since the methodology is strictly

21   deterministic, it can be readily parallelized by dividing the image into a plurality of tasks, which can

22   be executed by a plurality of processors in parallel. There is no need to take a step of ensuring that

23   positions of sample points are not correlated, which is generally necessary if a methodology based

24   on random or pseudo-random numbers is to be implemented for processing in parallel.

1   Generally, a computer graphics system that makes use of low-discrepancy sampling in

2   determination of sample points will perform better than a computer graphics system that makes use

3   of random or pseudo-random sampling, but the performance may degrade to that of a system that

4   makes use of random or pseudo-random sampling in higher dimensions. By providing that the

5   computer graphics system performs dependent splitting by replication, the superior convergence of

6   low-dimensional low-discrepancy sampling can be exploited with the effect that the overall

7   integrand becomes smoother, resulting in better convergence than with stratified random or pseudo-

8   random sampling. Since the computer graphics system also makes use of dependent trajectory

9   sampling by means of infinite low discrepancy sequences, consistent adaptive sampling of, for

10 example, light sources, can also be performed.

11 In addition, it will be appreciated that, although the computer graphics system has been

12 described as making use of sample points generated using generalized scrambled and/or unscrambled

13 Hammersley and Halton sequences, it will be appreciated that generally any (t,m,s)-net or (t,s)-

14 sequence can be used.

15 <u>Code Segment 1</u>

16 The following is a code fragment in the C++ programming language for generating the

17 positions of the jittered subpixel sample points x_i

```
18          unsigned short Period, *Sigma;
19          void InitSigma(int n)
20          {
21          unsigned short Inverse, Digit, Bits;
22          Period = 1 << n;
23          Sigma = new unsigned short [Period];
24          for (unsigned short i = 0; i < Period; i++)
25          {
26              Digit = Period
27              Inverse = 0;
28              for (bits = i; bits; bits >>= 1)
```

-27-

```
1        {
2          Digit >>= 1;
3          if (Bits & 1)
4            inverse += Digit;
5          }
6          Sigma[i] = Inverse;
7          }
8        }
9        void SampleSubpixel(unsigned int *i, double *x, double *y, int sx, int sy)
10       {
11         int j = sx & (Period - 1);
12         int k = sy & (Period - 1);
13         *i = j * Period + Sigma[k]
14         *x = (double) sx + (double) Sigma[k] / (double) Period;
15         *y = (double) sy + (double) Sigma[j] / (double) Period;
16       }
```

Code Segment 2

The following is a code fragment in the C++ programming language for generating a ray tree

```
19       class Ray
20       {
21         int i; //current instance of low discrepancy vector
22         int d; //current integral dimension in ray tree
23         ...
24       }
25       void Shade (Ray& Ray)
26       {
27         Ray next_ray;
28         int i = ray.i;
29         int d = ray.d
```

```
1        ...

2        for (int j = 0; j < M; j++)

3        {

4            ...

5            // ray set up for recursion
```

$$y = \left(\Phi_{b_d}\left(i,\sigma_{b_d}\right) \oplus \frac{j}{M}, \ldots, \Phi_{b_{d+\Delta d-1}}\left(i,\sigma_{b_{d+\Delta d-1}}\right) \oplus \Phi_{b_{d+\Delta d-2}}\left(i,\sigma_{b_{d+\Delta d-2}}\right) \right)$$

```
7            next_ray = SetUpRay(y);  // determine ray parameters by y

8            next_ray.i = i + j;  // decorrelation by generalization

9            next_ray.d = d + Δd;  //dimension allocation

10           Shade(next_ray);

11               ...

12           }

13       }
```

It will be appreciated that a system in accordance with the invention can be constructed in whole or in part from special purpose hardware or a general purpose computer system, or any combination thereof, any portion of which may be controlled by a suitable program. Any program may in whole or in part comprise part of or be stored on the system in a conventional manner, or it may in whole or in part be provided in to the system over a network or other mechanism for transferring information in a conventional manner. In addition, it will be appreciated that the system may be operated and/or otherwise controlled by means of information provided by an operator using operator input elements (not shown) which may be connected directly to the system or which may transfer the information to the system over a network or other mechanism for transferring information in a conventional manner.

The foregoing description has been limited to a specific embodiment of this invention. It will be apparent, however, that various variations and modifications may be made to the invention, with the attainment of some or all of the advantages of the invention. It is the object of the appended claims to cover these and such other variations and modifications as come within the true spirit and scope of the invention.